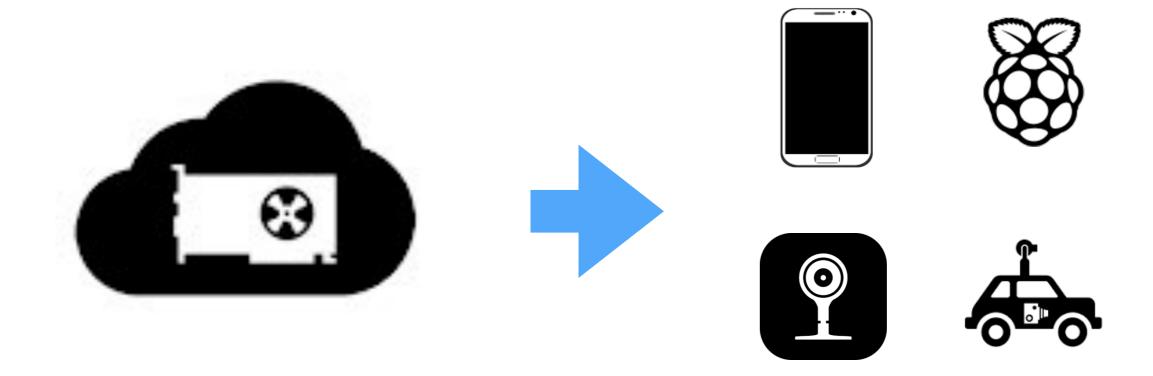
Deep Learning from a Mobile* Perspective

Yangqing Jia Research Scientist, Facebook

me@daggerfs.com

* mobile, on-device, embedded, IoT, anything without the cloud.

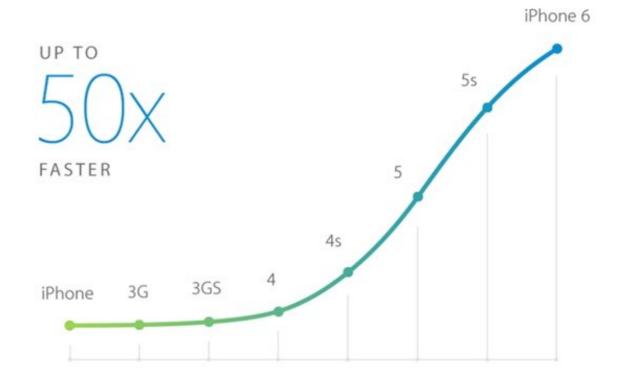
Why Embedded Systems?



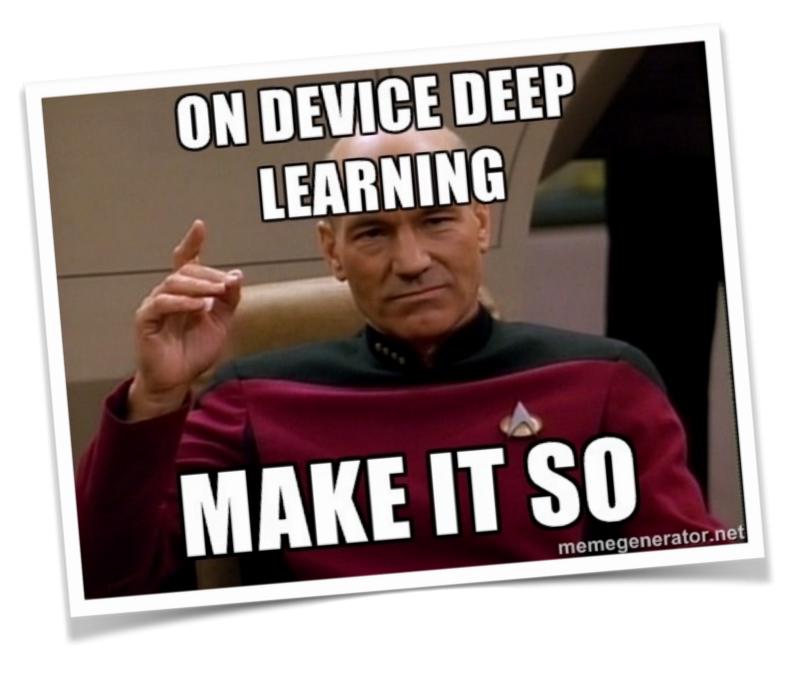
Expanding the capability of deep learning to a wide range of applications

Warp Speed

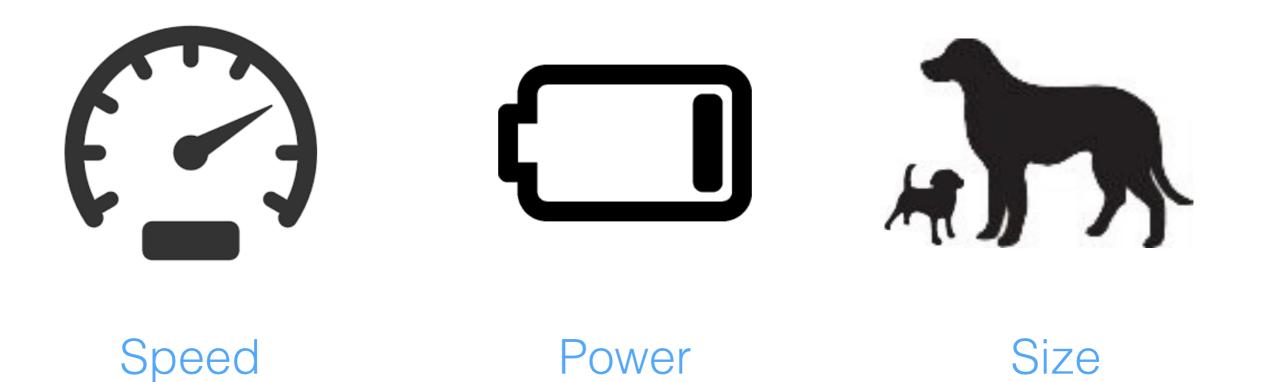
- Mobile is getting good!
 - Processor power
 - System architecture
 - Programming easiness
 - More applications

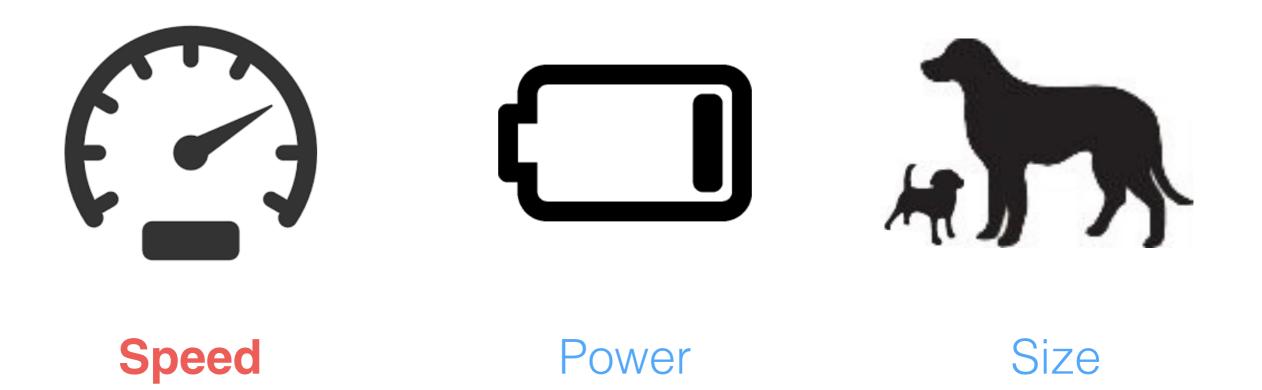


* Apple's processor growth over the years



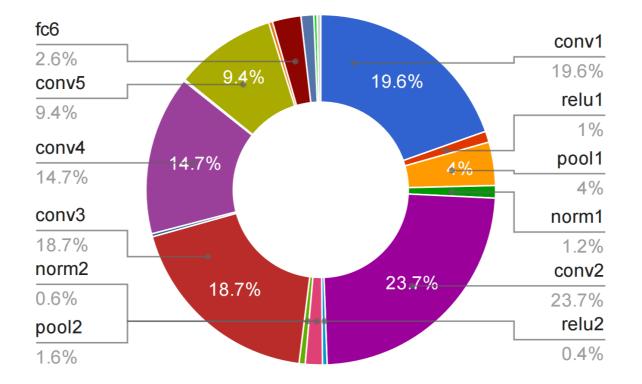
But wait, captain...





Speed

• Why Gemm/Conv is at the heart of Deep Learning



CPU Forward Time Distribution

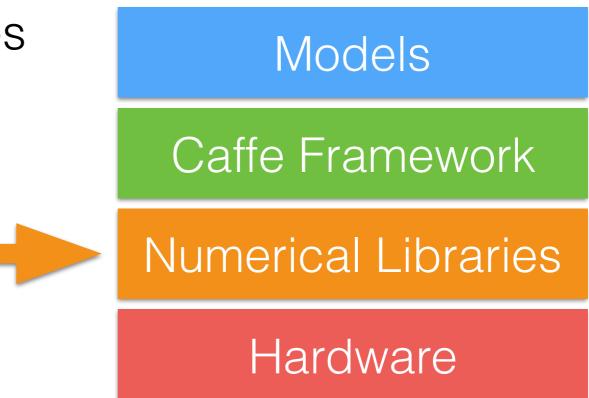
Source: UC Berkeley Thesis, Jia 2014

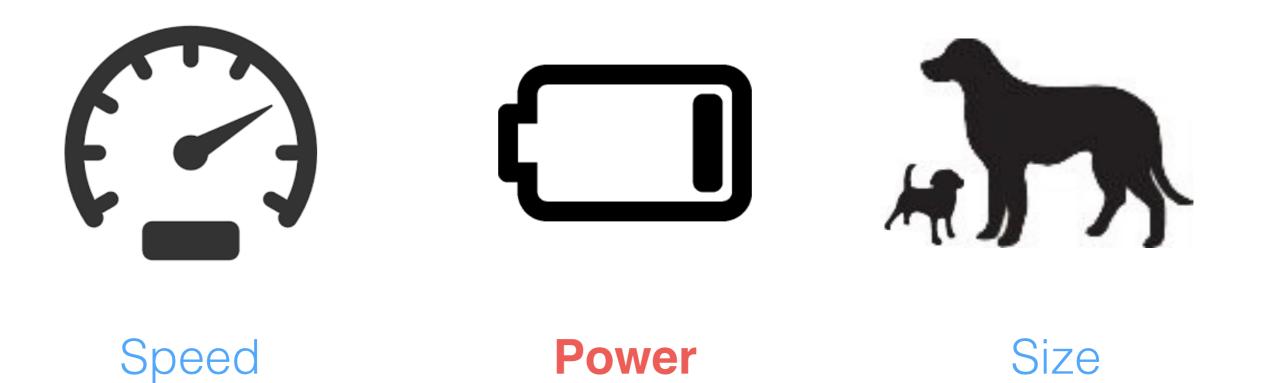
• Also, read Pete Warden's blog for more details

Towards a Ludicrous Speed

 We need better libraries for numerical optimization!
 CuDNN MKL Eigen

. . . .





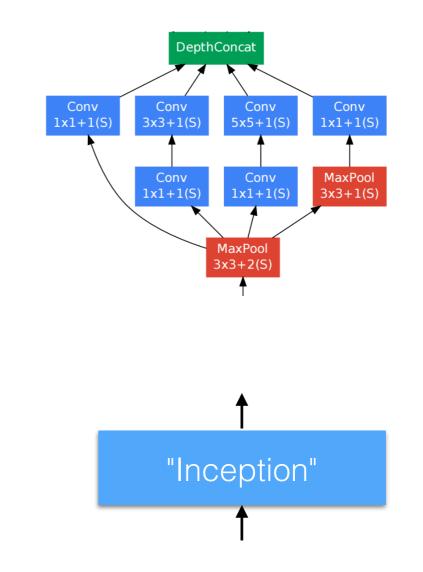
DNN Demands, Battery Suffers

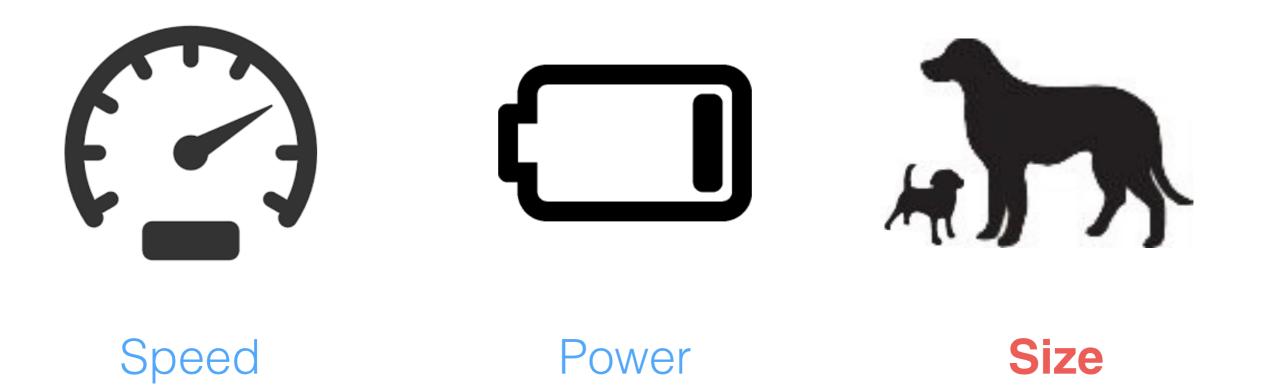
- DNNs are expensive Inception: 3 billion flops.
- Batteries don't last that long
 ~4 hrs (optimistic est.)
- Does not play well with thermo Burst in computation and intense use of cores



The Dilemma between Modularity and Efficiency

- It's all about balance
- SW: Fine-grained vs Coarse operators RTC / JIT?
- HW: Efficiency vs Programmability FPGA / ASIC?





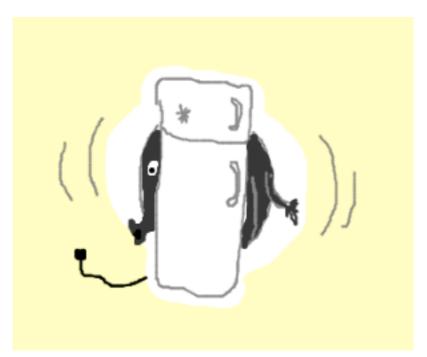
Putting Elephant in the Fridge

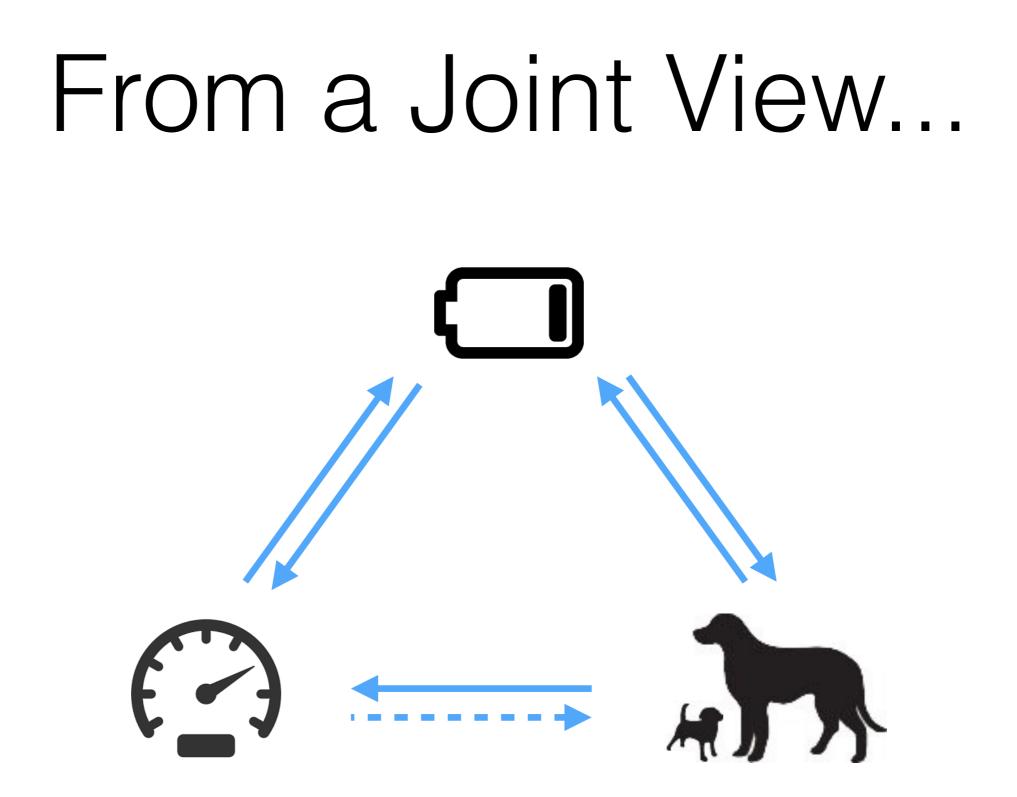
• DNN models are often very large

- We are getting better AlexNet: 240MB; Inception: 6MB
- But things are still wildly big for embedded Storage, bandwidth and memory limits

There are (Potential) Ways

- Model Compression Compress the model but keep (approx.) its math
- Better Model Designs Inception, separable convolutions, etc.
- Quantization
 Float -> float16, int8, custom format...
- Distillation [Hinton 2015]
 Train small models to mimic big ones





"To boldly go where no one has gone before."